

A finite cell approach for discretization of random fields

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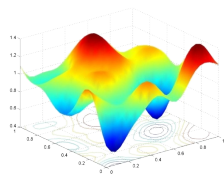
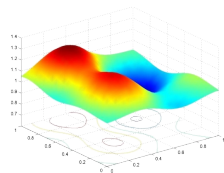
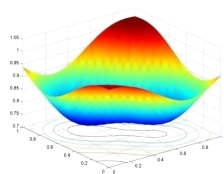


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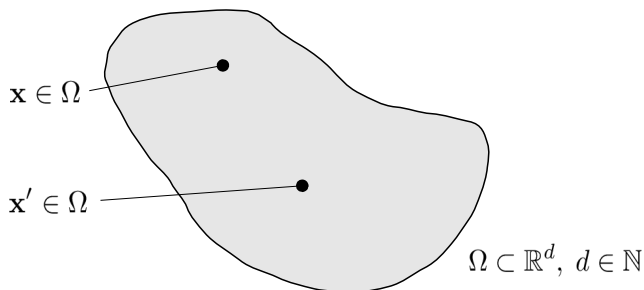
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Application of random fields - Examples:

- soil properties in geotechnical engineering
- groundwater heights
- rainfall



- 1 Introduction
 - Random fields
 - Karhunen-Loève (KL) expansion
- 2 Finite cell based KL-approximation
 - Higher order FEM discretization
 - Finite cell based discretization
- 3 Numerical Studies
 - 2D-Example
- 4 Summary



- random field (RF): $H(\mathbf{x})$
- *Gaussian* random field - completely described by:
 - mean function $\mu(\mathbf{x})$
 - covariance function $\text{Cov}(\mathbf{x}, \mathbf{x}') = \sigma(\mathbf{x}) \cdot \sigma(\mathbf{x}') \cdot \rho(\mathbf{x}, \mathbf{x}')$
 - $\sigma(\mathbf{x})$: standard deviation function
 - $\rho(\mathbf{x}, \mathbf{x}')$: correlation coefficient function

Random field discretization

Number of random variables (RVs) in a random field

- theoretically: *infinite* number of RVs (∞)
 - for each $\mathbf{x} \in \Omega$, $H(\mathbf{x})$ represents a RV
- discretized RF: *finite* number of RVs (M)

$$H(\mathbf{x}) \xrightarrow{\text{discretization}} \hat{H}(\mathbf{x}) \quad (1)$$

Categories of RF-discretization methods

- point discretization methods
- averaging discretization methods
- **series expansion methods**
 - Karhunen-Loève (KL) expansion
 - EOLE method

Karhunen-Loève (KL) expansion

(truncated) KL-expansion

$$H(\mathbf{x}) = \mu(\mathbf{x}) + \sum_{i=1}^{\infty M} \sqrt{\lambda_i} \varphi_i(\mathbf{x}) \xi_i \quad (2)$$

- λ_i : (M largest) eigenvalues of the covariance kernel
- φ_i : eigenfunctions of the covariance kernel
 - orthonormal: $\int_{\Omega} \varphi_i(\mathbf{x}) \varphi_j(\mathbf{x}) \, d\mathbf{x} = \delta_{ij}$
- ξ_i : uncorrelated standard normal RVs
 - orthonormal: $E[\xi_i \xi_j] = \delta_{ij}$

Integral eigenvalue problem

$$\int_{\mathbf{x}' \in \Omega} \varphi_i(\mathbf{x}') \text{Cov}(\mathbf{x}, \mathbf{x}') \, d\mathbf{x}' = \lambda_i \varphi_i(\mathbf{x}) \quad (3)$$

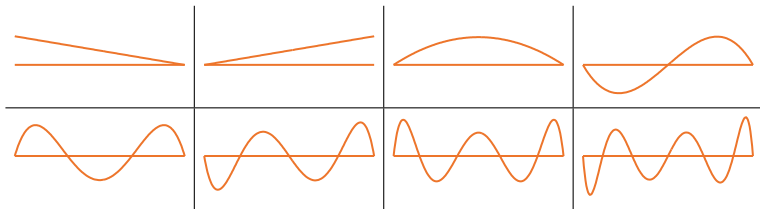
Approximation of the KL-eigenfunctions

Integral eigenvalue problem (KL-expansion)

$$\int_{\mathbf{x}' \in \Omega} \varphi_i(\mathbf{x}') \text{Cov}(\mathbf{x}, \mathbf{x}') d\mathbf{x}' = \lambda_i \varphi_i(\mathbf{x}) \quad (4)$$

Approximation of the eigenfunctions

$$\hat{\varphi}_i(\mathbf{x}) = \sum_{n=1}^N d_n^i N_n(\mathbf{x}) = \mathbf{d}_i^T \mathbf{N}(\mathbf{x}); \quad N_n(\mathbf{x}) \in L^2(\Omega) \quad (5)$$



Matrix eigenvalue problem

Matrix eigenvalue problem

$$\mathbf{B}d_i = \hat{\lambda}_i \mathbf{M}d_i \quad (6)$$

where

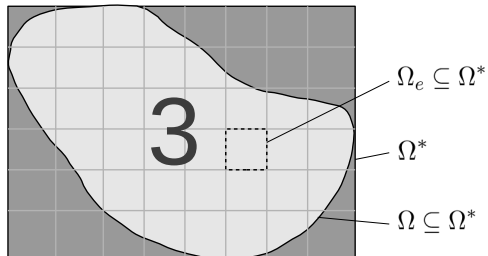
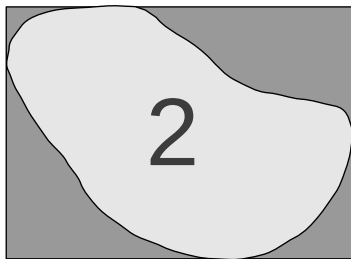
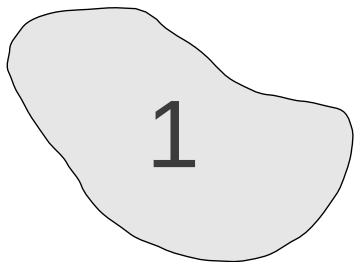
$$B_{kn} = \int_{\mathbf{x} \in \Omega} N_k(\mathbf{x}) \int_{\mathbf{x}' \in \Omega} N_n(\mathbf{x}') \text{Cov}(\mathbf{x}, \mathbf{x}') d\mathbf{x}' d\mathbf{x} \quad (7)$$

$$M_{ij} = \int_{\mathbf{x} \in \Omega} N_i(\mathbf{x}) N_j(\mathbf{x}) d\mathbf{x} \quad (8)$$

Approximated truncated KL-expansion

$$\hat{H}(\mathbf{x}) = \mu(\mathbf{x}) + \sum_{i=1}^M \sqrt{\hat{\lambda}_i} \hat{\varphi}_i(\mathbf{x}) \xi_i \quad (9)$$

Finite cell - basic idea & notation



- global shape functions:
 $N_i \in L^2(\Omega^*)$
- indicator function:

$$\alpha(\mathbf{x}) = \begin{cases} 1 & \forall \mathbf{x} \in \Omega \\ 0 & \forall \mathbf{x} \in \Omega^* \setminus \Omega \end{cases}$$

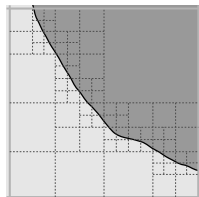
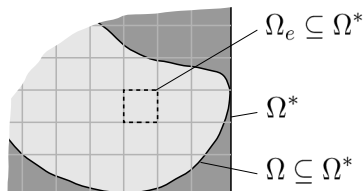
Matrix eigenvalue problem

$$\mathbf{B} \mathbf{d}_i = \hat{\lambda}_i \mathbf{M} \mathbf{d}_i \quad (10)$$

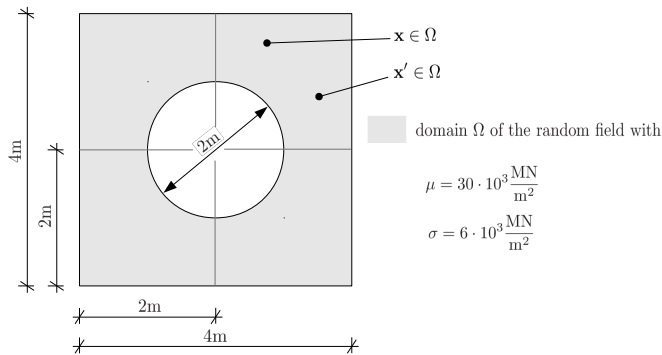
where

$$B_{kn} = \int_{\mathbf{x} \in \Omega^*} \alpha(\mathbf{x}) N_k(\mathbf{x}) \int_{\mathbf{x}' \in \Omega^*} \alpha(\mathbf{x}') N_n(\mathbf{x}') \text{Cov}(\mathbf{x}, \mathbf{x}') \, \mathrm{d}\mathbf{x}' \, \mathrm{d}\mathbf{x} \quad (11)$$

$$M_{ij} = \int_{\mathbf{x} \in \Omega^*} \alpha(\mathbf{x}) N_i(\mathbf{x}) N_j(\mathbf{x}) \, \mathrm{d}\mathbf{x} \quad (12)$$



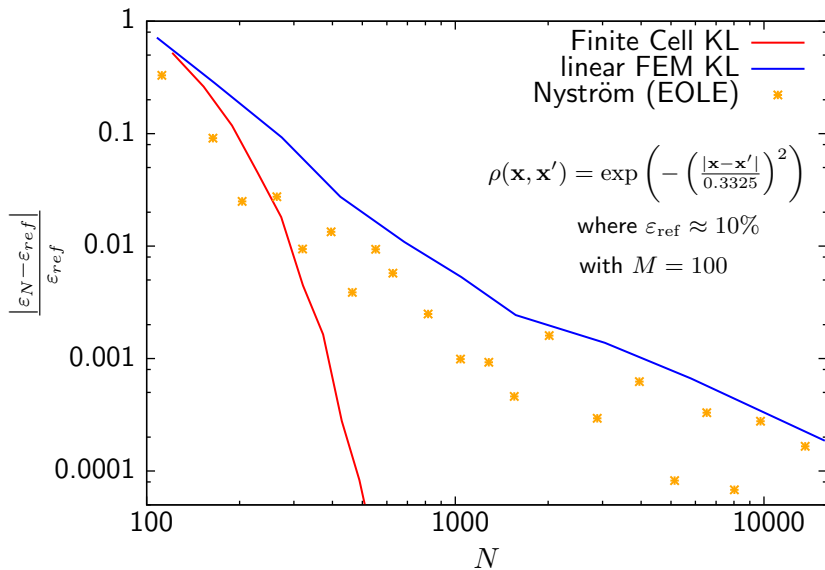
Example of a plate with a hole



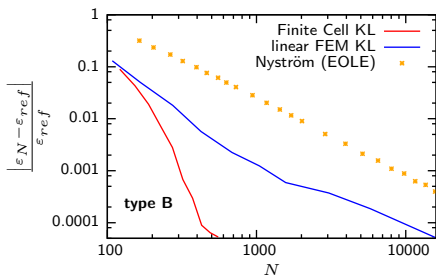
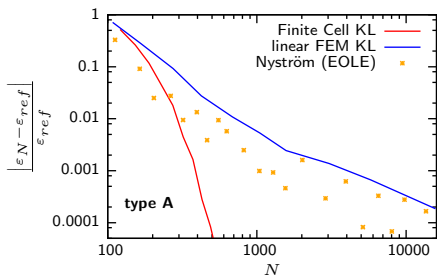
Investigated discretization methods

- **Finite Cell KL** (4 cells; increase in polynomial order)
- **linear FEM KL** (linear shape functions; increase number of elements)
- **Nyström (EOLE)** (increase number of points)

1st study: convergence of relative error w.r.t. N



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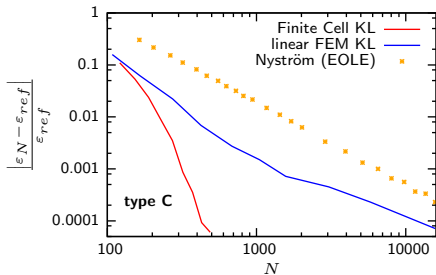


A: $\rho(\mathbf{x}, \mathbf{x}') = \exp\left(-\left(\frac{|\mathbf{x}-\mathbf{x}'|}{0.3325}\right)^2\right)$

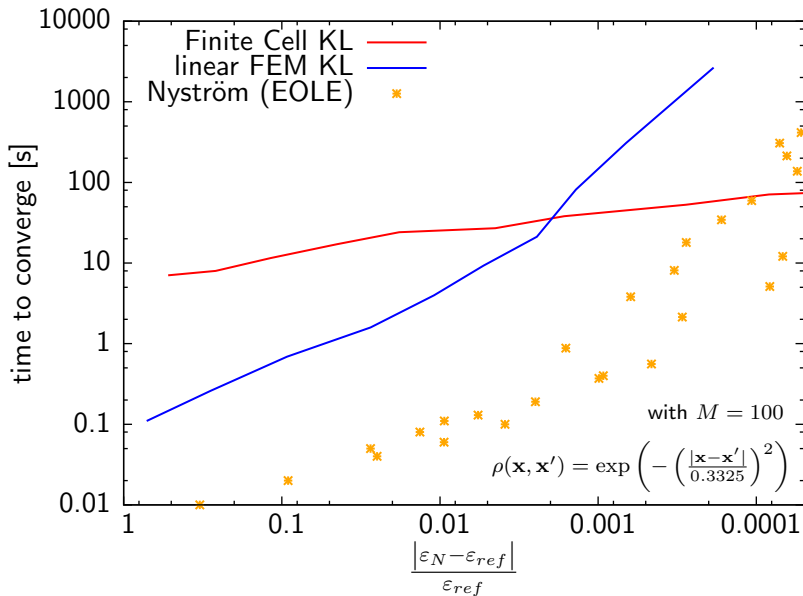
B: $\rho(\mathbf{x}, \mathbf{x}') = \exp\left(-\frac{|\mathbf{x}-\mathbf{x}'|}{1.08}\right)$

C: $\rho(\mathbf{x}, \mathbf{x}') = \frac{1}{1+\left(\frac{|\mathbf{x}-\mathbf{x}'|}{0.725}\right)^{1.2}}$

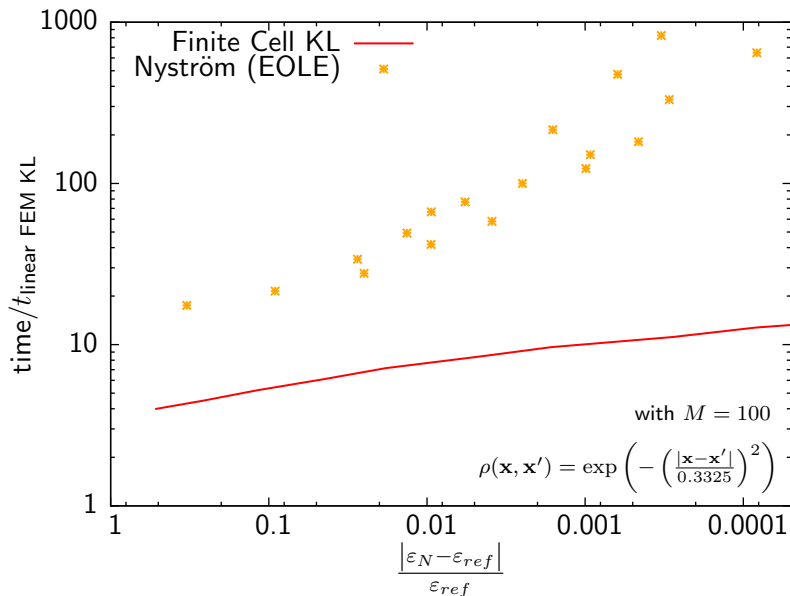
where $\varepsilon_{ref} \approx 10\%$



2nd study: time needed to obtain a random field approx.



3rd study: time needed to compute a realization



Summary and Conclusion

FC-KL - Cons

- Computationally very **expensive to solve** (*especially in 3D*)
 - FC-KL: double integral over covariance function
 - EOLE: just $N \times N$ covariance function (and Lanczos methods)
- Quite **difficult to implement** (*compared to EOLE*)
 - higher order shape functions
 - (double) integration of non-continuous non-smooth functions
- **Numerical stability** of eigenvalue problem

FC-KL - Pros

- **Fast convergence** against optimal representation
- Realization computationally **cheap to evaluate**
 - Efficient assembly of FE stiffness matrices
- **Simple mesh**